

PART 1

1.) Estimate Autoregressive Integrated Moving Average (ARIMA(p,d,q)) model for spot and future – determine the most appropriated order for p, d, and q using SBIC given the maximum lag equals 4.

```
. tsset time
      time variable:  time, 1 to 795
              delta:  1 unit

. g y = spot *set spot value as y in order to use do file easier"
. do "/var/folders/qx/mw3zz03s545c66rmml_hrzfc0000gn/T//SD16322.000000"

. forvalues p = 1(1)4{
  2. forvalues q = 1(1)4{
  3.             display " estimate arima`p'1`q'"
  4.             qui arima y, arima(`p',1,`q') nolog
  5.             est store arima`p'1`q'
  6. }
  7. }
estimate arima111
estimate arima112
estimate arima113
estimate arima114
estimate arima211
estimate arima212
estimate arima213
estimate arima214
estimate arima311
estimate arima312
estimate arima313
estimate arima314
estimate arima411
estimate arima412
estimate arima413
estimate arima414

.
end of do-file

. est table arima11*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable		arima111	arima112	arima113	arima114
Y					
	_cons	-.10471744	-.10468155	-.10474014	-.10717398
ARMA					
	ar				
	L1.	.43536547*	.32884791	-.21609401	.59490084
	ma				
	L1.	-.35351081	-.25287121	.29330109	-.52209152
	L2.		.02661686	.06275321	-.00167449
	L3.			.07084457**	.03360908
	L4.				-.07201064*
sigma					
	_cons	8.1399985***	8.1383248***	8.1283318***	8.1204174***
Statistics					
	N	794	794	794	794
	ll	-2791.4965	-2791.3458	-2790.3816	-2789.5823
	chi2	18.312657	16.97941	13.218602	28.319443
	aic	5590.993	5592.6915	5592.7633	5593.1646
	bic	5609.7013	5616.077	5620.8258	5625.9042

legend: * p<.1; ** p<.05; *** p<.01

```
. est table arima21*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable	arima211	arima212	arima213	arima214
y				
_cons	-.10453586	-.10767542	-.10609456	-.10747347
ARMA				
ar				
L1.	.30210361	1.3247307***	.46523894***	1.3085163***
L2.	.02452891	-.92390998***	-.83397031***	-.92906275***
ma				
L1.	-.22494448	-1.3022249***	-.39216993***	-1.2433923***
L2.		.94779741***	.82697929***	.88679293***
L3.			.09595475***	.05677728
L4.				.00003472
sigma				
_cons	8.1387104***	8.077039***	8.1149358***	8.0663508***
Statistics				
N	794	794	794	794
ll	-2791.3835	-2785.5252	-2789.1057	-2784.5179
chi2	16.085901	3656.4671	241.72156	4791.7034
aic	5592.7671	5583.0504	5592.2113	5585.0358
bic	5616.1525	5611.1129	5624.9509	5622.4524

legend: * p<.1; ** p<.05; *** p<.01

. est table arima31*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima311	arima312	arima313	arima314
y				
_cons	-.10486338	-.10606433	-.10806841	-.10765208
ARMA				
ar				
L1.	-.56058829	.56814277***	.66749387**	.58444954
L2.	.08569307*	-.89762847***	-.06995188	.02671133
L3.	.06374492**	.09752317***	-.59846871*	-.67269654
ma				
L1.	.63856195	-.49161073***	-.61040004*	-.51894361
L2.		.88783708***	.0556601	-.03501301
L3.			.65798111**	.71899187
L4.				.02307553
sigma				
_cons	8.1302906***	8.1136284***	8.0655263***	8.0640794***
Statistics				
N	794	794	794	794
ll	-2790.5308	-2788.9397	-2784.4054	-2784.2793
chi2	21.334447	289.47751	1722.9017	1770.3817
aic	5593.0617	5591.8793	5584.8108	5586.5587
bic	5621.1242	5624.6189	5622.2274	5628.6524

legend: * p<.1; ** p<.05; *** p<.01

. est table arima41*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima411	arima412	arima413	arima414
y				
_cons	-.10781888	-.10743716	-.10771017	-.10767653
ARMA				
ar				
L1.	.6019371*	1.3677396***	.61533738	.83865029
L2.	-.00196743	-1.0093746***	.0094391	-.13015924
L3.	.02357655	.05751154	-.67660585	-.67766115
L4.	-.074501**	-.001562	.02303694	.17029234
ma				

L1.		-.52565375	-1.302657***	-.54898831	-.77133898
L2.			.96445922***	-.01948257	.10786573
L3.				.72230601	.72482037
L4.					-.15743316

sigma					
_cons		8.1178063***	8.066311***	8.0639684***	8.0633837***

Statistics					
N		794	794	794	794
ll		-2789.3297	-2784.5185	-2784.2686	-2784.2263
chi2		30.742143	5310.9545	1882.2706	2721.9204
aic		5592.6594	5585.037	5586.5371	5588.4525
bic		5625.399	5622.4537	5628.6309	5635.2234

legend: * p<.1; ** p<.05; *** p<.01					

for the spot arima(1,1,1) is the most appropriate model since it has the lowest BIC

```
. drop y
. gen y = future *set spot value as y in order to use do file easier"

. do "/Users/jharit/Desktop/arima do file.do"

. forvalues p = 1(1)4{
  2. forvalues q = 1(1)4{
  3.     display " estimate arima`p'1`q'"
  4.     qui arima y, arima(`p',1,`q') nolog
  5.     est store arima`p'1`q'
  6. }
  7. }
estimate arima111
estimate arima112
estimate arima113
estimate arima114
estimate arima211
estimate arima212
estimate arima213
estimate arima214
estimate arima311
estimate arima312
estimate arima313
estimate arima314
estimate arima411
estimate arima412
estimate arima413
estimate arima414

.
end of do-file

. est table arima11*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable		arima111	arima112	arima113	arima114

y					
_cons		-.11424575	-.11408115	-.1139183	-.11370744

ARMA					
ar					
L1.		-.36850497	.30224152	.07471074	-.26441554
ma					
L1.		.33119538	-.33885872	-.11128313	.23060131
L2.			.04456173	.0345267	.02935071
L3.				.03319847	.04074154
L4.					.03843721

sigma					
_cons		9.5427695***	9.5390832***	9.5360585***	9.532552***

Statistics					

N	794	794	794	794
ll	-2917.724	-2917.4236	-2917.166	-2916.8767
chi2	3.3320137	3.4714053	5.5726249	7.6343946
aic	5843.4479	5844.8473	5846.3321	5847.7535
bic	5862.1563	5868.2327	5874.3946	5880.493

 legend: * p<.1; ** p<.05; *** p<.01

. est table arima21*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima211	arima212	arima213	arima214
y				
_cons	-.11399264	-.11299023	-.11518755	-.114745
ARMA				
ar				
L1.	.26006773	1.1957056***	.39680958***	.29779715***
L2.	.03927244	-.91443628***	-.83492246***	-.89274399***
ma				
L1.	-.29537225	-1.2256874***	-.43709178***	-.33588403***
L2.		.97207404***	.86844397***	.9442071***
L3.			.02125732	.00060118
L4.				.05136959*
sigma				
_cons	9.5398841***	9.450655***	9.4808276***	9.473342***
Statistics				
N	794	794	794	794
ll	-2917.4896	-2910.3454	-2912.6217	-2912.0146
chi2	2.8388146	6763.0586	566.77521	690.44183
aic	5844.9792	5832.6907	5839.2433	5840.0292
bic	5868.3646	5860.7532	5871.9829	5877.4458

 legend: * p<.1; ** p<.05; *** p<.01

. est table arima31*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima311	arima312	arima313	arima314
y				
_cons	-.11395314	-.11434855	-.11482471	-.11489771
ARMA				
ar				
L1.	.06416111	1.3204613***	-.5619286***	.01972589
L2.	.02920804	-.90163025***	-.42025289***	-.77499053***
L3.	.02294871	-.05791248**	-.83389309***	-.25249026
ma				
L1.	-.10003354	-1.3735497	.51793334	-.05826621
L2.		1.0000008	.40777623	.81652449***
L3.			.88983828	.26256084
L4.				.05469308*
sigma				
_cons	9.5380895***	9.4185571	9.4577901	9.4722994***
Statistics				
N	794	794	794	794
ll	-2917.3418	-2909.7635	-2911.5297	-2911.9403
chi2	3.8262796	32923.58	7392.6354	480.87614
aic	5846.6836	5833.5271	5839.0595	5841.8806
bic	5874.7461	5866.2666	5876.4761	5883.9743

 legend: * p<.1; ** p<.05; *** p<.01

. est table arima41*, star(0.1 0.05 0.01) stat(N ll chi2 aic bic)

Variable	arima411	arima412	arima413	arima414
y				

	_cons	-.11336618	-.11461152	-.11590868	-.11376705

ARMA					
	ar				
	L1.	-.26142231	.3181852***	1.7676402***	-.41944632***
	L2.	.01694481	-.85050807***	-1.5054804*	.36718541***
	L3.	.03124469	.00130972	.41843749	-.35522863***
	L4.	.0307881	.04773204	.0005867	-.86090936***
	ma				
	L1.	.22521179	-.35750674***	-1.8167695***	.38743681***
	L2.		.90090477***	1.6207452*	-.30501342***
	L3.			-.49952575	.42560261***
	L4.				.8658178***

sigma					
	_cons	9.5350812***	9.4739792***	9.4436415***	9.4058783***

Statistics					
	N	794	794	794	794
	ll	-2917.0927	-2912.0811	-2909.6096	-2906.4848
	chi2	5.4882963	661.06102	4207.3335	932.73412
	aic	5848.1853	5840.1622	5837.2193	5832.9696
	bic	5880.9249	5877.5789	5879.313	5879.7404

legend: * p<.1; ** p<.05; *** p<.01					

for the future arima(2,1,2) is the most appropriate model since it has the lowest BIC

2.) Make dynamic forecast for period time = 796 to 800.

dynamic forecast from 796 to 800

```
. arima spot, arima(1,1,1) nolog
```

ARIMA regression

Sample: 2 - 795	Number of obs	=	794
	Wald chi2(2)	=	18.31
Log likelihood = -2791.496	Prob > chi2	=	0.0001

	D.spot	Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]

spot						
	_cons	-.1047174	.3525647	-0.30	0.766	-.7957316 .5862967

ARMA						
	ar					
	L1.	.4353655	.2343856	1.86	0.063	-.0240219 .8947528
	ma					
	L1.	-.3535108	.242659	-1.46	0.145	-.8291138 .1220922

	/sigma	8.139999	.1507405	54.00	0.000	7.844553 8.435444

Note: The test of the variance against zero is one sided, and the two-sided confidence interval is truncated at zero.

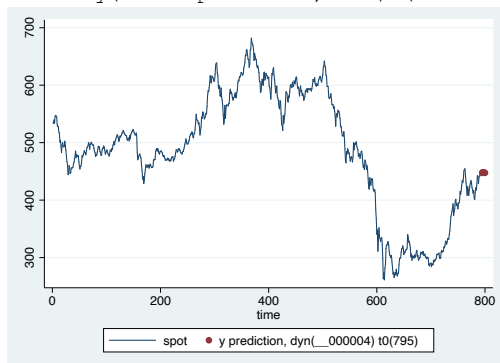
```
. set obs 800
```

number of observations (_N) was 795, now 800

```
. replace time=_n
(5 real changes made)
```

```
. predict spothat, y dynamic(.) t0(795)
Note: beginning dynamic predictions in period
(794 missing values generated)
```

```
. twoway(line spot time ,sort) (scatter spothat time,sort)
```



```
. arima future, arima(2,1,2) nolog
```

ARIMA regression

Sample: 2 - 795

Number of obs = 794

Wald chi2(4) = 6763.06

Log likelihood = -2910.345

Prob > chi2 = 0.0000

D.future		Coef.	OPG Std. Err.	z	P> z	[95% Conf. Interval]	

future							
_cons		-.1129902	.3496789	-0.32	0.747	-.7983483	.5723678

ARMA							
ar							
L1.		1.195706	.0266001	44.95	0.000	1.14357	1.247841
L2.		-.9144363	.0262076	-34.89	0.000	-.9658023	-.8630703
ma							
L1.		-1.225687	.0174538	-70.22	0.000	-1.259896	-1.191479
L2.		.972074	.0174447	55.72	0.000	.937883	1.006265

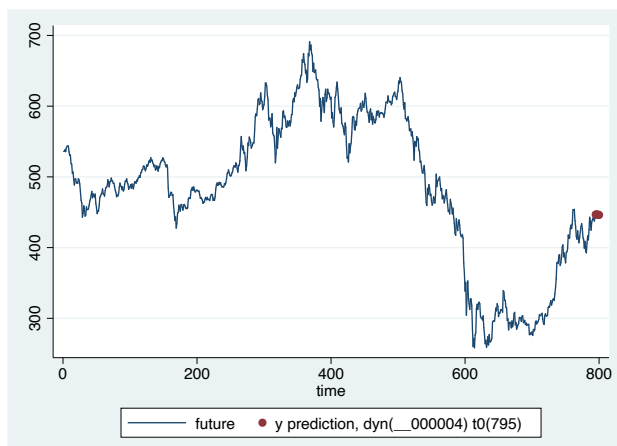
/sigma		9.450655	.1666509	56.71	0.000	9.124025	9.777285

Note: The test of the variance against zero is one sided, and the two-sided confidence interval is truncated at zero.

```
. predict futurehat, y dynamic(.) t0(795)
```

Note: beginning dynamic predictions in period 6.
(794 missing values generated)

```
. twoway(line future time ,sort) (scatter futurehat time,sort)
```



Part2

3.)Estimate model (1) using OLS by employing future return (rfuture) as dependent variable and spot return (rspot) as explanatory variable, and determine whether ARCH-effect significantly occurs.

```
gen rspot = spot-1.spot
(6 missing values generated)
```

```
. gen rfuture = future - 1.future
(6 missing values generated)
```

```
. replace rspot = rspot/spot
(793 real changes made)
```

```
. replace rfuture = rfuture/future
(792 real changes made)
```

```
. reg rfuture rspot
```

Source	SS	df	MS	Number of obs	=	794
Model	.3430572	1	.3430572	F(1, 792)	=	6161.14
Residual	.044099197	792	.000055681	Prob > F	=	0.0000
Total	.387156397	793	.000488217	R-squared	=	0.8861
				Adj R-squared	=	0.8860
				Root MSE	=	.00746

rfuture	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
rspot	1.103312	.0140562	78.49	0.000	1.07572 1.130904
_cons	-.0000445	.0002649	-0.17	0.866	-.0005645 .0004754

```
. estat archlm
```

LM test for autoregressive conditional heteroskedasticity (ARCH)

lags(p)	chi2	df	Prob > chi2
1	34.112	1	0.0000

H0: no ARCH effects vs. H1: ARCH(p) disturbance

Reject H0: then, we have arch effect in the model

3.)Estimate GARCH(p,q) for future return (rfuture) using spot return (rspot) as explanatory variable for mean equation (model (1) and (2)) – determine the most appropriated order p and q for variance equation using SBIC given the maximum lag equals to 2.

```
do "/Users/jharit/Desktop/garch do file.do"
```

```
. forvalues p = 1(1)2{
  2. forvalues q = 1(1)2{
  3.           display " estimate garch`p'`q'"
  4.           qui arch rfuture rspot, arch(1/`p') garch(1/`q') nolog
  5.           est store garch`p'`q'
  6. }
  7. }
estimate garch11
estimate garch12
estimate garch21
estimate garch22
```

```
.
end of do-file
```

```
. est table garch* , star(0.1 0.05 0.01) stat(N ll chi2 aic bic)
```

Variable		garch11	garch12	garch21	garch22
rfuture					
	rspot	1.0645681***	1.064238***	1.0682047***	1.0680005***
	_cons	-.00004419	-.00005155	-.00006558	-.00006707
ARCH					
	arch				
	L1.	.08508817***	.11662409***	.22294495***	.22541563***
	L2.			-.16183182***	-.18077526***
	garch				
	L1.	.89766172***	.51492192	.92480421***	1.1004611***
	L2.		.34282779		-.15477774
	_cons	9.520e-07***	1.374e-06**	7.402e-07***	5.126e-07**
Statistics					
	N	794	794	794	794
	ll	2853.3906	2854.746	2858.4282	2858.8165
	chi2	13442.953	12918.945	12438.335	12846.434
	aic	-5696.7811	-5697.492	-5704.8564	-5703.633
	bic	-5673.3957	-5669.4295	-5676.7939	-5670.8934

legend: * p<.1; ** p<.05; *** p<.01

Garch(1,2) is the most appropriated one since it has the lowest bic

5.) From (4), predict the variance of future return (rfuture).

```
arch rfuture rspot, arch(1/2) garch(1) nolog

ARCH family regression

Sample: 2 - 795                                Number of obs   =          794
Distribution: Gaussian                          Wald chi2(1)    =       12438.33
Log likelihood = 2858.428                      Prob > chi2     =          0.0000
```

		OPG		z	P> z	[95% Conf. Interval]	
rfuture		Coef.	Std. Err.				
rfuture							
	rspot	1.068205	.009578	111.53	0.000	1.049432	1.086977
	_cons	-.0000656	.0001916	-0.34	0.732	-.000441	.0003099
ARCH							
	arch						
	L1.	.2229449	.0531304	4.20	0.000	.1188113	.3270786
	L2.	-.1618318	.0538085	-3.01	0.003	-.2672946	-.0563691
	garch						
	L1.	.9248042	.0140053	66.03	0.000	.8973544	.9522541
	_cons	7.40e-07	2.32e-07	3.19	0.001	2.85e-07	1.20e-06

```
. predict sigma2 , v
. line sigma2 time
```

